





# On Transaction-Based Metrics as a Proxy for Profitability of Financial Asset Recommendations

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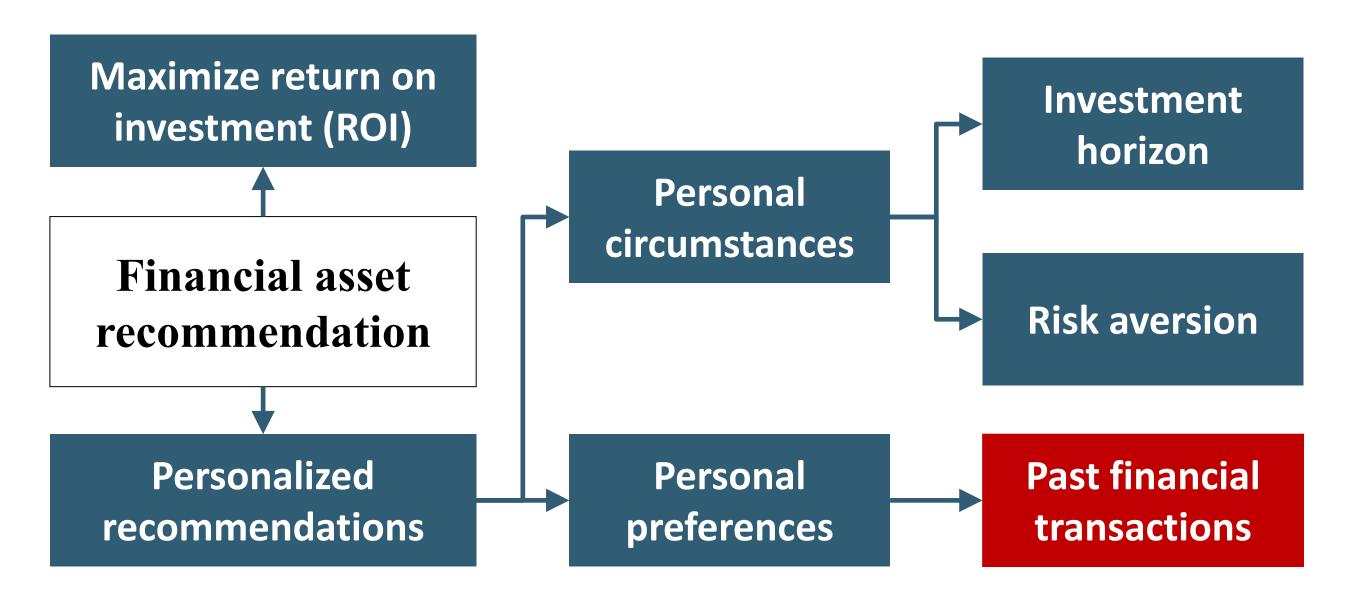
Overview

Financial Asset Recommender (FAR) systems use models trained on past data to suggest investment assets to customers. The effectiveness of FAR models can be evaluated in two ways: (a) measuring the money customers could obtain if they followed the recommendations (profitability-based) and (b) quantifying the ability of models to predict future customer investments (transaction-based). In this work we compare these strategies to determine which should be the primary metric for evaluating FAR systems.



Paper

## 1. Task and Motivation



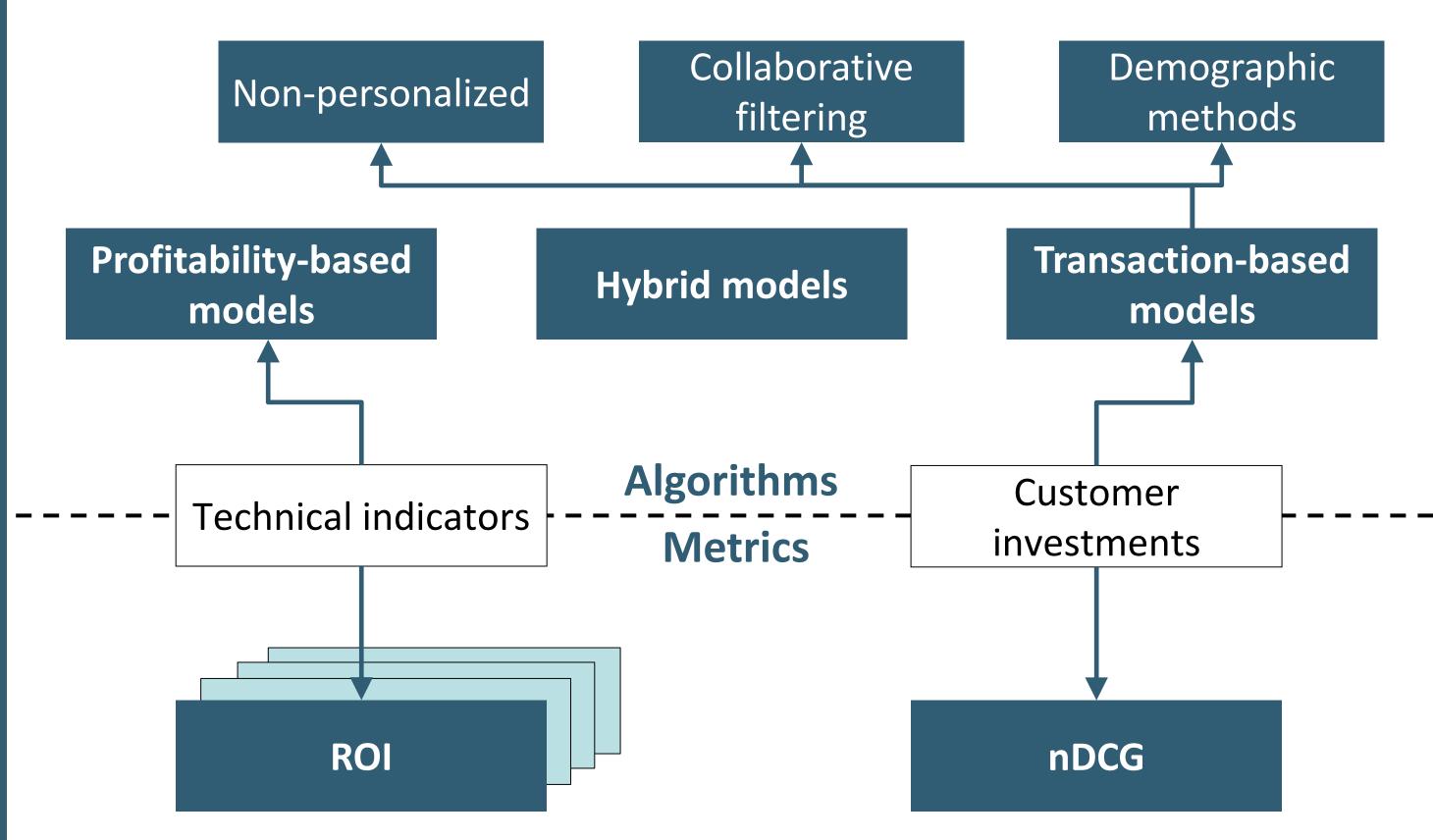
#### Transaction-based metrics

- > calculate similarity between a set of recommendations and what the customer actually invested in (financial transactions) as a quality metric over the produced recommendations
- > would be expected to be positively correlated with return on investment if customers are effective investors.
- > would be a better aggregate measure of a FAR system, as they are inherently personalized if this is true

**Research** Are transaction-based metrics to be positively correlated with return on investment and if not, why not?

# 2. Experiment Design

 We train a wide range of FAR systems that consider different types of input data to evaluate



• We compare how these models perform if the customer invests for 6 months under both transaction and profitability metrics

## 3. Dataset

- Snapshot of the Greek market covering a range of different securities: stocks, bonds, mutual funds and other banking products for the period between January 2018 and March 2021
  - > Evaluation performed for 29 time points, spaced 2 weeks apart
  - > Starting from 1st July 2019
  - Data previous to each time point as training, six months after as test

Market data		Customer data			
Property	Value	Property	Value		
Unique assets	5,371	Unique customers	52,390		
Assets with investments	2,025	Transactions	313,004		
Price data points	1,768,128	Acquisitions	269,031		
Average return (by assets, whole period)	23.67%	Average return (by customers, whole period)	18.41%		
% profitable assets	53.08%	% customers with profits	58.00%		

# 4. Results

 RQ1: Are transaction-based and profitability-based metrics interchangeable when evaluating financial asset recommendation systems?

		Transaction- Based				fitability- based
Data source	Category	Algorithm	nDCG@10	ROI@10		
None		Random	0.0223	0.0118	_	
Technical indicators		SVR	0.0041	0.1212		Best ROI /
	Regression	LightGBM	0.0599	0.1423		Worst nDCG
		Random forest	0.0570	0.0583		VVOISCIIDCG
Past investments	Non-personalized	Popularity	<u>0.3374</u>	-0.0628		
		LightGCN	0.3081	-0.0643		
	Collaborative	ARM	0.2687	-0.0647		Best nDCG /
	filtering	MF	0.0812	-0.0460		Worst ROI
		UB kNN	0.1428	-0.0344		vvoist koi
	Demographic	CPS	0.3003	-0.0544		
ام نسمایی ا		Hybrid-nDCG	0.2454	-0.0466		
Hybrid		Hybrid-regression	0.0220	0.0382		
Market average				0.1026		

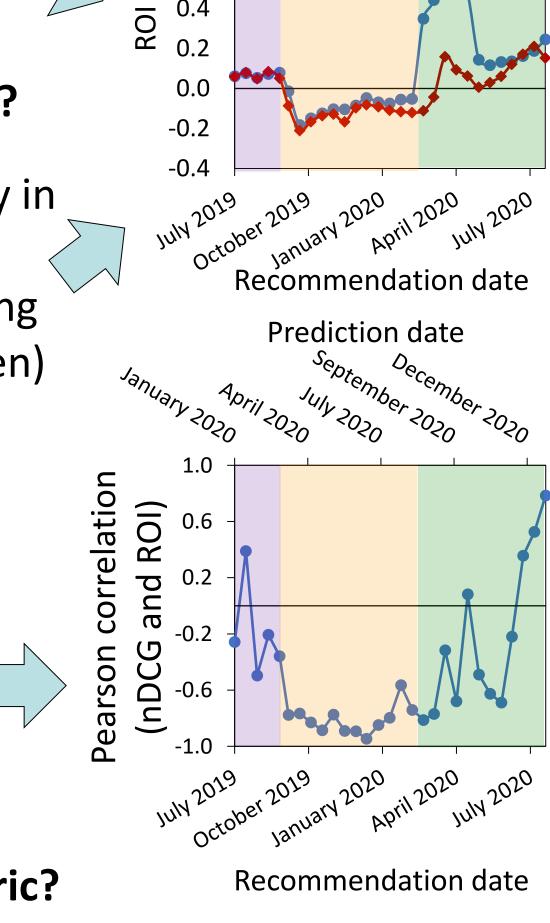
 No, profitability-based metric and transaction-based metrics are not correlated – models can perform well under transaction-based metrics but still loose money

• RQ2: What are the main factors that influence transaction-based metrics?

Are our customers poor investors?
On average the customers in this dataset underperform the market; this seems to be exacerbated by high volatility in the market



- We can see this unusual fall in profitability in late 2019 (COVID-19 impact, in yellow)
- Also, a spike in profitability when predicting for early 2020 (COVID-19 recovery, in green)
- There were far fewer profitable assets during the downturn, followed by many profitable assets during the market rally
- During these downturn, volatile times, transactions and profitability become negatively correlated



0.8

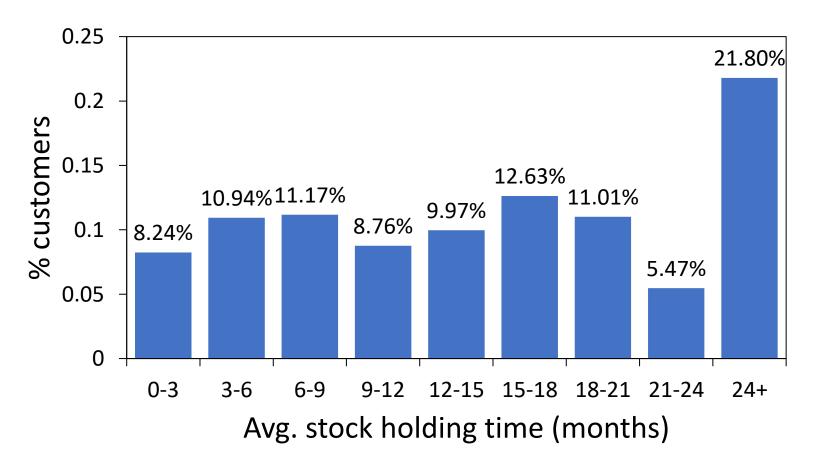
0.6

Prediction date

Customers

Is ROI after 6 months a bad profitability metric?

Only for a subset of the customers, we can see that there is a wide spread in terms of holding time amongst the customers, indicating that a range of prediction time horizons are needed



## 5. Conclusion

Multiple factors such as variable investment horizons, market volatility and the ability of the customers to navigate the market make transaction-based metrics risky as an evaluation metric for FAR systems.



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